



Econometric Perspective of the Oil Consumption and Economic Growth Relation in India

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Abstract

This research aims to study the existence and direction of the short-run or long-run relationship between per capita oil consumption and gross domestic product in India. The data from 1965 to 2015 had analyzed by employing the vector error correction model. For the verification of the same result, a standard Granger causality test had performed. The study results have suggested the existence of a long-run relationship and show the direction of changes in gross domestic product cause changes in oil consumption. As a policy implication, economic growth can be considered a policy variable to improve India's oil resources.

Keywords: Oil consumption; Economic growth; Error correction model; Granger causality test

Introduction

Over the past two decades, the rapid growth of the Indian economy has reflected its growing demand for petroleum products. As of 2018, India was the world's third-largest user of 5.2 million barrels of oil per day, exceeded only by the world's two largest economies, the US and China. Consumption has risen almost 1.7 times since 2008 and has exceeded the figures for Japan. India also reported more substantial growth than China post-2010, particularly between 2015 and 2016, when consumption increased by 8 per cent-10 percent per year [1,2]. India expected to overtake China in terms of oil consumption by the mid-2020s. Table 1 shows the production and consumption of oil per capita in India.

At the same time, the massive demand for commodities in India can hardly satisfy domestic oil production. Oil production in the last decade, between 35 and 37mn tonnes per year, was relatively constant. In addition to the adverse effects of geopolitical instability and volatile petroleum

prices, the production of Indian oil by 2024 could increase significantly. However, recent data show that crude oil production otherwise decreased continuously since 2012 and declined to 34.2 million tonnes in 2019, the lowest amount since 2010 [1,2]. The aging of petroleum fields in the country hampers crude oil production. India's territory also makes it difficult for oil and gas to explore, and the requisite skills exceed the abilities of local government companies. The usual assumption is that it is an electric source or a diesel source for vehicles concerning the use of petroleum. Petroleum products have a much wider variety of uses for the manufacture in almost every kind of daily life of asphalt, highway gasoline, various plastics, chemical, and synthetic materials. There is no distinction in this regard for India, fuel like gasoline, engine spirits, and turbine aviation, used commonly in petroleum products: bitumen (used for road building), grades and lubricants, and oil coke. However, fuel is still the bulk of oil consumption, with 39.1 percent of total oil consumption dropping into high-speed diesel and the engine's spirit (14.2 percent) and liquified fuel gas (12.2 percent).

Year	Oil Production	Oil Consumption
2005	353.359	1239.12
2006	359.097	1283.69
2007	357.604	1360.99
2008	365.479	1405.68
2009	362.313	1461.22
2010	388.235	1468.09
2011	398.219	1520.09
2012	390.303	1598.24
2013	385.006	1594.97
2014	373.18	1625.55
2015	364.884	1739.88

Table 1: Oil production and oil consumption in India.

Due to its value in all life aspects, it is not surprising that economists consider their consumption as a useful economic measure of success. As a leading indicator, oil consumption and GDP now have templates for developed and developing

economies. Indian data shows that a similar decline in petroleum consumption occurred at the start of the GDP growth deceleration in Q1 2018. Figure 1 shows the growth pattern between oil and gross domestic product in India.

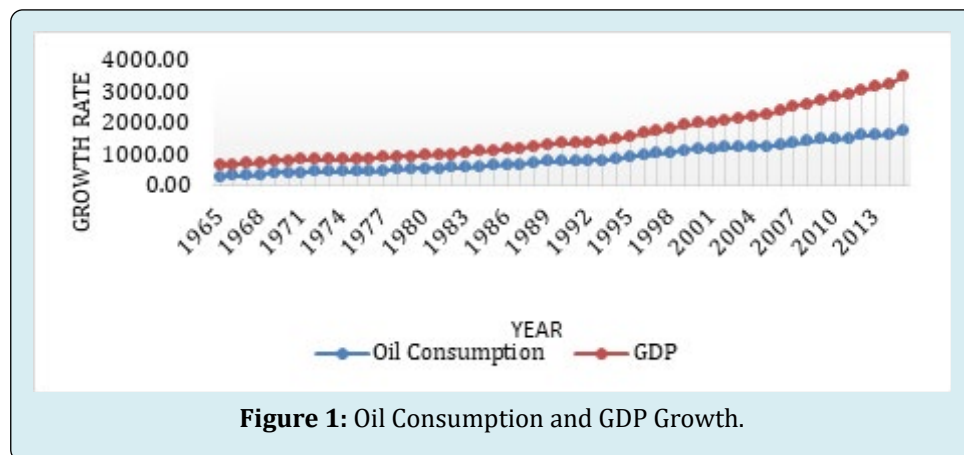


Figure 1: Oil Consumption and GDP Growth.

India's strategy for tackling its overreliance on oil imports is mainly based on two actions to promote domestic development and the focus on renewable energy sources. In its oil consumption policies, the Indian government had more than helped, enabling up to 100 percent FDI in many petroleum and gas sectors. The program had designed to promote exploration and development of petroleum and natural gas and decrease dependence on imports, such as HELP (hydrocarbon Exploration and Licensing Policy) and OALP (Open Acreage Licensing Program).

So, looking at the importance of oil consumption in economic development, this research paper has attempted to analyze the interaction between oil production and India's gross domestic product. The study's core problem discussed

in this section is the second section, Data, and Methodology, followed by the third section, Results, and Discussion. Based on the given results, the Conclusion and some Policy Implications had suggested in the last section.

Data and Methodology

Data

To find the relationship between per capita oil consumption and per capita gross domestic product in India, the annual data took for 1965-66 to 2015-16 (the latest available). The data of per capita oil consumption and per capita gross domestic product (Constant 2010 US\$) from www.data.worldbank.org. A three-stage procedure was

employed to know the relationship, and the study results had calculated with the help of statistical application EViews (IHS Global Inc., USA).

Unit Root Test

As a first step, Augmented Dickey-Fuller (ADF) performed to know the stationarity of time series data and a null hypothesis set, i.e., series is non-stationary [3]. So, the following intercept took into consideration:

$$\Delta Y_t = \alpha_1 + \alpha_{oc} OC_t + \mu_t \quad (1)$$

Where Y_t , OC_t and μ_t are Gross Domestic Product, Oil Consumption and μ_t the error term.

Johansen Co-Integration

Co-Integration test helps in verifying the null hypothesis H_0 : There is no co-integration [4]. The two parameters of the Johansen co-integration tests are the Trace test and maximum eigenvalue test.

$$LR(r_0, n) = -T \sum_{i=r_0+1}^n \ln(1 - \lambda_i) \quad (2)$$

$$LR(r_0 + r_0 + 1) = T \ln(1 - \lambda_{r_0+1}) \quad (3)$$

Vector Error Correction Model (VECM)

As the third stage will enable exploring both the short

Results and Discussions

Statistics	GDP	Oil Consumption
Mean	720.6437	860.8132
Median	575.5015	771.3959
Maximum	1751.664	1739.876
Minimum	345.4216	295.4104
Std. Dev.	387.1662	419.5817
Skewness	1.117942	0.480350
Kurtosis	3.140107	1.943548
Jarque-Bera	10.66497	4.332951
Probability	0.004832	0.114581
Sum	36752.83	43901.48
Sum Sq. Dev.	7494884.	8802441.
Observations	51	51

Table 2: Descriptive Statistics of Included Variables.

and long-run dynamics of co-integrated sequence [5]. The conventional ECM for co-integrated series is:

$$\Delta LOC_t = \alpha_o + \sum_{i=1}^n \alpha_i \Delta LOC_{t-i} + \sum_{i=0}^n \delta_i \Delta LGDP_{t-i} + \varphi z_{t-1} + \mu_t \quad (4)$$

$$\Delta LGDP_t = \beta_0 + \sum_{i=1}^n \beta_i \Delta LGDP_{t-i} + \sum_{i=0}^n \delta_i \Delta LOC_{t-i} + \varphi z_{t-1} + \mu_t \quad (5)$$

Where, z is the ECT (Error Correction Term), and it is the OLS (Ordinary Least Square) residual from the following long-run co-integrating regression: $LOC_t = \alpha_o + \alpha_1 LGDP_t + \mu_t$ and $LGDP_t = \beta_0 + \beta_1 LGDP_t + \mu_t$ are defined as $z_{t-1} = ECT_{t-1} = LOC_{t-1} - \alpha_o - \alpha_1 LGDP_{t-1}$. To verify the results of VECM, a standard Granger Causality Test is also employed, which validates the direction of causality flow from one to the other variables and vice versa [6]. The model of the Granger causality test is as follows:

$$\Delta LOC_t = \alpha + \sum_{i=1}^k \beta_i \Delta LOC_{t-i} + \sum_{i=1}^k \gamma_i \Delta LGDP_{t-i} + \mu \quad (6)$$

$$\Delta LGDP_t = \alpha + \sum_{i=1}^k \beta_i \Delta LGDP_{t-i} + \sum_{i=1}^k \psi_i \Delta LOC_{t-i} + \mu \quad (7)$$

Whereas α , β , ψ , and γ are parameters to be estimated and μ represent the serial error terms, LOC_t and $LGDP_t$ are defined observation for the t periods; Δ is the differential operator; k refers to the number of lags; α , β , ψ , and γ all are the parameters of the estimation.

Unit Root Test

Firstly, Augmented Dickey-Fuller (ADF) had performed to test for a unit root in the level series and the lag length based on the SI criterion, which was an automatic selection. As the results indicated in Table 3, the p-value of both series

(Oil Consumption and Gross Domestic Product) was less than 5 percent at the first difference, including intercept in the test equation. So, the null hypothesis H_0 : Series has a unit root e.g., series is non-stationary was rejected.

Variables at Level	ADF Test (P-Values)*	First Difference (P- Values)*
LOC	-6.652446	0.0000
LGDP	-6.521700	0.0000

Table 3: Unit-Root Test (Augmented Dickey-Fuller).

(Note: *indicates significant at the 1 percent level, LGDP- Log Gross Domestic Product, LOC- Log Oil Consumption)

The statistics of both tests were stationary; however, before performing the Johansen co-integration test, there was

a need to determine an optimal number of lags or lag length selection criteria by using Vector Autoregression (VAR).

Lag	LogL	LR	FPE	AIC	SIC	HQ
0	11.145	NA	0.0023	-0.3896	-0.3106	-0.3599
-1	202.35	358.09*	8.06e-07*	-8.3550*	-8.1191*	-8.2660*
2	204.16	3.239	8.86E-07	-8.2629	-7.8681	-8.1147

Table 4: Result of Lag Length Criteria.

(Note: *Indicates lag order, LR- sequential modified LR test statistics (each test at 5% level, FPE- Final prediction error, AIC- Akaike's information criterion, SIC- Schwarz information criterion, HQ- Hannanquinn information criterion)

In Table 4, different values of lag 0 to 2 had shown by log L, LR, FPE, AIC, SC, and HQ. However, lag (1) had a more considerable negative value. So, (1) was determined as lag length selection criteria and chosen to perform the co-integration and other relevant tests.

Johansen Co-integration Test

The Johansen co-integration study was useful in understanding how two or more stochastic time series co-integrate. The initial Johansen test is a confirmation of the null hypothesis H_0 : There is no co-integration for both test statistics.

Hypothesized No. of CE(s)	Eigenvalue	Trace Statistic	0.05* Critical Value	Prob.**
None	0.390726	24.7321	15.49471	0.0015
At most 1	0.019571	0.948707	3.841466	0.33

Table 5: Unrestricted Cointegration Rank Test (Trace and Eigenvalue).

*denotes rejection of the hypothesis at the 0.05 level, **MacKinnon-Haug-Michelis (1999) p-values

Table 5 shows the Johansen co-integration test results with the help of the trace test & eigenvalue. The values shown by hypothesized None, trace test & eigenvalue test accepted the null hypothesis, i.e., H_0 : There is no co-integration at the 5 percent level of significance. These results suggest that there is a short-term relationship between oil consumption and gross domestic product. On the other side, the hypothesized findings Most one could not be rejected by the null hypothesis because the p-value was more significant than the 5 percent level. These results indicate a single co-

integration and an error term in the model. Therefore, the Johansen co-integration test confirmed that both the oil consumption and gross domestic product series have a short-run equilibrium relationship.

Vector Error Correction Model

As given the Johansen co-integration test outcome, by running the Vector Error Correction Model (Table 6) to analyze how long-run deviations are associated.

Variable		
	-0.05074	0.051944
T-statistics	-2.73938	3.48724
P-value	0.009	0.0012
	-0.00804	-0.01368
T-statistics	-0.05674	-0.12004
P-value	0.955	0.905
	-0.24483	-0.07148
T-statistics	-1.81426	-0.65866
P-value	0.0768	0.5137
	0.379498	-0.14168
T-statistics	2.047952	-0.95072
P-value	0.0469	0.3472
	0.391251	-0.0885
T-statistics	2.067546	-0.5815
P-value	0.0449	0.564
C	0.019374	0.042981
T-statistics	1.725972	4.761176
P-value	0.0917	0

Table 6: Results of vector error correction model.

The LOC had a negative value (-0.05074) and a statistically significant p-value (0.009), which satisfies both conditions for verifying the long-run relationships between variables (and), i.e., in the long run, the changes in Gross Domestic Product cause changes in Oil Consumption. The negative value (-0.134863) suggested that if there were a departure in one direction, the correction would have to be pulled back in another direction. To ensure that equilibrium had retained, about 5 percent (-0.05074) of departure from

long-run equilibrium was corrected for each period. and refer to the coefficient for the lag values of the target variable LOC. The and prove statistically insignificant, which indicates the absence of short-run causality running from GDP to OC based on VECM estimates. As the results suggested by Table 7, the null hypothesis could not be accepted because the p-value (0.3990) for this Chi-square statistic was more than 5 percent. So, there was an absence of long-run causality from LGDP to LOC.

Test statistic	Value	DoF*	P-Value**	Decision
F-statistic	3.50856	(2, 42)	0.399	Accept
Chi-square	7.01711	2	0.299	Accept

Table 7: Wald test.

(Note: *Degree of Freedom, ** denotes rejection of the hypothesis at the 0.05 level)

There was also a need to verify that there must be no serial correlation between the variables. To verify, by selecting the *Breusch-Godfrey Serial Correlation LM Test* in residual diagnostic by determining two lags, as suggested by lag length selection criteria in Table 4. The null hypothesis

was H_0 : There is no co-integration and the results in Table 8 suggested that based on Prob. Chi-Square value (0.4152) the null hypothesis cannot be rejected because the p-value was more significant than 5 percent. So, there was no evidence of serial correlation, which was a good sign for the model.

F-statistic	Obs* R-squared	Prob. F (2,40)	Prob. Chi-Square (2) *	Decision
0.760249	1.757779	0.4742	0.4152	Accept

Table 8: Breusch-Godfrey serial correlation LM test.

(Note: * denotes the hypothesis at the 0.05 level)

Granger Causality Test

To confirm the estimates of the VECM Granger causality test performed. This test helps verify whether there is a long-run relationship between LOC and LGDP or the absence of a short-run relationship. The results of Table 9, based on F-statistics, suggested that “LGDP does not Granger Cause

LOC” at the 5 percent significance level. This finding confirms the previous results of VECM (Table 6 & 7) that there is an absence of short-run causality from at the 5 percent significance level. So, in the long run, the changes in Gross Domestic Product cause changes in Gross Oil Consumption and not in the short run in India.

Null Hypothesis:	Observation	F-Statistic	Prob.	Decision
LGDP does not Granger Cause LOC	49	2.06433	0.1390	Accept
LOC does not Granger Cause LGDP		0.10227	0.9030	Accept

Table 9: Pairwise Granger Causality Tests.

(Note: *indicates the rejection of null hypotheses at the 5 percent significant level, LGDP- Log Gross Domestic Product, LOC- Log Oil Consumption

Conclusion

This research attempts to remove the dust from this controversial issue by investigating the dynamic interaction between per capita oil consumption and gross domestic product in India from 1965 to 2015. Interestingly, the statistical results by employing the Vector Error Correction Model (VECM) indicate a long-run equilibrium relationship. The results imply that changes in gross domestic product cause changes in oil consumption and the absence of short-run relationship and same verified through the Granger causality test. The study’s findings contribute to the growth of the oil industry by revealing the importance of strategic policy formulation and the implications of economic growth. As a policy implication, economic growth can be considered a policy variable to improve India’s oil resources. India still fails to achieve its goals concerning renewable energy. The target of 175 GW of renewed power by 2022 seems rather ambitious, given that by December 2019, the total capacity was only around 86 GW, according to the Center for Science and the Environment (CSE). The renewable energy sources’ growth has slowed down to several factors, including slow auction production, higher cost, and project hazards. Furthermore, the combination of a slowing economy and diversification barriers indicates a continuous rise in reliance on oil imports. India has fewer instruments at its hands than before in the current situation. It may not be an

option between stimulating growth and oil independence.

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